

# Simulation of Grain-Straw Separation by Discrete Element Modeling with Bendable Straw Particles

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## Abstract

The combine harvester owes its name to the integration of the whole chain of grain harvesting steps in one machine. Running these interdependent processes simultaneously requires insight in the influence of crop factors and adjustments on the individual processes and the consequences for downstream processes. This paper introduces changes to the Discrete Element Method (DEM) in order to be suitable for the simulation of grain-straw separation, which is one of the most critical processes in the combine harvester. Segmented bendable straw particles have been constructed in the DEMeter++ simulation environment and their physical properties have been calibrated with realistic straw properties. The use of these particles for modeling separation has been validated by reconstructing an existing separation experiment by Beck (1992) in DEMeter++ and comparing the simulation result with the experiment. Once validated, the practical use of the simulation framework to assess the sensitivity of separation to crop properties is illustrated.

*Keywords:*

DEM, Bendable cylindrical particles, Separation, Straw, Grain, Combine harvester

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## 1. Introduction

Grain and straw have a different shape and density. Grain can therefore be separated from straw by accelerating the mixture. In the separation section of a combine harvester, the grain kernels which have been released from the ears through threshing have to be expelled from the straw layer. The necessary acceleration is induced by the action of oscillating walkers or centrifugation. Obviously, as the combine harvester is a continuous-flow machine, the residence time of the grain-straw mixture in the threshing and separation unit limits the time available for the grain to migrate through the straw layer to the walker's surface. Grain kernels that are threshed but are still lodged in the straw layer when it leaves the machine are called separation losses. Modern combine harvesters separate more than 100 tons of this grain and straw mixture per hour, meanwhile, the acceptable grain loss over all sub-processes is typically not higher than 1%. The separation and cleaning losses usually have the biggest share. Designing a new separation section and optimizing its settings is not straightforward due to the complex interaction between the machine and the crop. In addition, for an optimal operation of existing separation sections, quantitative insight in the interference of crop, design and settings is required. An accurate model describing the separation process in a grain harvester would allow to (Kutzbach, 2003):

- Reduce the test expenditure
- Improve the understanding of the fundamental relationships
- Make a targeted choice on future improvements

- 42      • Simulate the influences of different parameters
- 43      • Estimate possible performance increases

44    A modeling framework for separation should be able to describe the parti-  
45    cle interactions adequately and to include particle properties involved. The  
46    models which have been reported in literature (summarized e.g. in Kutzbach  
47    (2003)) fit well to the experimental data, but their empirical nature typically  
48    limits their practical relevance to specific crop and machine characteristics.  
49    Furthermore, the effect of crop properties on separation is mostly unclear as  
50    different crop property combinations can give the same end result in separa-  
51    tion. However, the importance of their influence on the separation process  
52    has been underlined by Hall and Husman (1981), Shandilya (1987) and Sri-  
53    vastava et al. (1990). As many of these properties are interdependent (Hall  
54    and Husman, 1981; Huisman, 1978) it is impossible to correlate individual  
55    crop properties to separation performance statistically. Moreover, the biolog-  
56    ical variability of the crop in the field is high and the strong effect of feedrate  
57    on separation will shade the effect of other factors. In order to have better  
58    control over the variables involved and to be able to monitor the process  
59    more accurately, laboratory experiments on separation have been carried out  
60    under known conditions of crop weight and properties. In this context an  
61    idealized experiment representative of separation on straw walkers, employed  
62    by Baader et al. (1969) has been repeated in several research works in the  
63    last decades, for instance by Huisman (1978) and Beck (1992). The setup  
64    consists of a straw layer contained in a vertically, sinusoidally oscillating box.  
65    After the straw has been agitated for a certain time, a layer of grain kernels is  
66    released at once on top of the oscillating straw layer. Grain passage through

67 the grating at the bottom of the box is recorded in function of time. The  
68 grain sinks through the straw layer, while it disperses laterally. This results  
69 in a sigmoidal curve of the separated fraction in function of time. The lower  
70 the area density of the straw layer, the faster the kernels can sink and the  
71 steeper the separation curve will be. Shandilya (1987) used a similar setup  
72 but with horizontal instead of vertical shaking.

73     So far little is known about the mechanical interactions between grain and  
74 straw particles during separation. This would yield fundamental insight into  
75 the influence of the properties of the particles involved on the performance.  
76 As this interaction takes place at the particle level, a modeling framework on  
77 this level should be set up. When modeling the behaviour of a collection of  
78 particles, like grain kernels, Discrete Element Modeling (DEM) is a logical  
79 choice as it allows to model the behaviour of each kernel through its inter-  
80 actions with the other kernels and the system elements. Also in agricultural  
81 processes, DEM is more and more used for simulating particulate processes  
82 as e.g. grain flow in silos (González-Montellano et al., 2011), fertilizer spread-  
83 ing (Tijskens et al., 2003; Van Liedekerke et al., 2009) or manure handling  
84 and land application equipment (Landry et al., 2006). In this way, the influ-  
85 ence of particle properties and boundary conditions can be assessed with a  
86 set of *in silico* experiments that can be run in parallel. However, simulating  
87 the straw particles with discrete elements is more challenging as straw has a  
88 large aspect ratio, resulting in a clear orientation. To study the alignment  
89 of straw particles and cutting blades in the chopper section of a combine,  
90 Kattenstroth et al. (2011) employed a Discrete Element Model with straw  
91 particles constructed with connected spheres. However, straw is bendable, a

92 property which is expected to have an important impact on the separation  
93 process. To the authors' knowledge, DEM simulation with bendable straw  
94 particles has not been reported in scientific literature to date.

95 Therefore, in this study a discrete element approach including bendable  
96 straw has been implemented in the DEMeter++ (Tijskens et al., 2003) soft-  
97 ware. Once the straw particles have been constructed, the crop properties  
98 can be easily defined and changed independently. This makes it relatively  
99 easy to study the sensitivity of the separation profile with reference to the  
100 individual crop properties. The Discrete Element Model with bendable straw  
101 particles is then used to simulate the stationary separation experiment de-  
102 scribed by Beck (1992). As the properties of the straw and grain kernels  
103 employed in the real-life experiment are not known, an experimental design  
104 of crop properties is set up and a simulation is run with each set of crop prop-  
105 erties. The validation then consists of a statistical comparison of simulated  
106 separation profiles to the reported experimental profiles. Once validated, the  
107 different separation profiles can be used to perform a sensitivity analysis of  
108 the crop properties on the separation rate.

## 109 **2. Discrete Element Modeling framework**

110 The Discrete Element Method (DEM), also called Discrete Element Mod-  
111 eling, is a numerical technique to model the motion of an assembly of particles  
112 which interact with each other through collisions. It was originally developed  
113 by Cundall and Strack (1979) for predicting the behaviour of soil grains and  
114 belongs to the group of "Particle Based Simulations". By applying DEM,  
115 the trajectory of each particle in a system can be obtained using a numerical

time integration scheme. At each time step, all forces acting on the particles like contact forces, body forces, etc. are summed. Newton's equations of motion are then integrated to obtain the velocity and position of each particle at the next time step (Tijskens et al., 2003). A DEM problem can be described mathematically as a system of non-linear differential ODE's formed by Newton's equations of translation (Eq. 1) and rotation (Eq. 2) for each individual particle  $i$ :

$$m_i \mathbf{a}_i = \mathbf{G}_i + \sum_c \mathbf{F}_{ci}, \quad (1)$$

$$\mathbf{I}_i \boldsymbol{\alpha}_i = \mathbf{H}_i + \sum_c \mathbf{r}_{ci} \times \mathbf{F}_{ci}, i = 1, \dots, N. \quad (2)$$

$\mathbf{a}_i$  and  $\boldsymbol{\alpha}_i$  are the translational and rotational acceleration of the  $i$ th particle and  $\mathbf{m}_i$  and  $\mathbf{I}_i$  are the mass and inertia tensor.  $\mathbf{G}_i$  and  $\mathbf{H}_i$  are the body force and moment that act on the  $i$ th particle. Additionally  $\mathbf{F}_{ci}$  is the contact force acting on the particle caused by the  $c$ th contact with a neighboring particle. Finally,  $\mathbf{r}_{ci}$  represents the position vector of that  $c$ -th contact with respect to the center of mass of the particle  $i$

The DEM algorithm of the grain-straw separation has following structure:

**System initialization.** The state of the system is completely determined at  $t=0$ . The initial dimensions of all particles and their initial positions and velocities are defined. Also the time discretization step ( $\Delta t$ ) is chosen.

**At each time step:**

- *Contact detection.* All current contacts between particles should be de-

139 tected. To find all these contacts in a computationally efficient way, the  
140 grid based contact detection algorithm proposed by Iwai et al. (1999)  
141 is used.

- 142 • *Contact forces.* For each actual contact between particle  $i$  and another  
143 particle  $j$  the contact force  $F_{ij}$  is calculated. Particles in DEM are con-  
144 sidered as rigid bodies with point contacts. Deformation during impact  
145 is simulated by allowing particles to overlap slightly. Consequently the  
146 contact force between two particles is related to this virtual overlap.  
147 The applied contact model can be divided into two parts: a normal  
148 and a tangential part. A linear viscoelastic model, namely a linear  
149 spring-damper, is used as normal contact force model (Tijskens et al.,  
150 2003), whereas the Werner-Haff model (Haff and Werner, 1986) is used  
151 as tangential contact force model (Haff and Werner, 1986).
- 152 • *Compilation of forces and moments.* The vectors of the forces working  
153 on each individual particle  $i$  are summed and also used to calculate  
154 moments. These resulting forces and moments are then employed in  
155 Newton's equations of motion (Eq.1 and 2).
- 156 • *Integration of the equations of motion.* Newton's equations of motion  
157 are integrated numerically to obtain the new positions and velocities  
158 of each individual particle  $i$  at time  $t + \Delta t$ . Numerical integration  
159 schemes are mostly kept very basic in DEM as the numerical errors of  
160 these schemes are much less important than the approximate nature of  
161 the contact force models. As for this reason the second order accurate  
162 Leap-frog time integration algorithm is used (Tijskens et al., 2003).

163 After the appropriate settings of the DEMeter++ program have been  
164 chosen, the particles are constructed. This is described in the next subsection.

### 165 *2.1. Grain particles in DEMeter++*

166 Grain kernels are approximated by spheres. This is a rough approxi-  
167 mation of real grain kernels, rather having an ellipsoid shape. More realistic  
168 grain kernels could be created by using composite particles made up of several  
169 overlapping spheres. However, the additional computational cost of compos-  
170 ite particles would have made it infeasible to simulate the experiment within  
171 a reasonable time. As characterizing the grain shape with only one dimen-  
172 sion led to a successful approximation of the separation curve in Gregory and  
173 Fedler (1986) and Shandilya (1987) it is expected that this assumption can  
174 be safely made if the right dimension of the grain is chosen as diameter of  
175 the spheres.

### 176 *2.2. Straw particles in DEMeter++*

177 A realistic straw stalk should be flexible in all directions, be extensible  
178 and compressible in the axial and radial direction and have frictional char-  
179 acteristics. To satisfy all these requirements segmented straw particles have  
180 been developed in DEMeter++. Each segment consists of a rigid hollow  
181 cylinder connected to its adjacent cylinders by spherical joints. At both ends  
182 of the straw particle an additional joint is placed, as can be seen in Fig.  
183 1. The joints are responsible for the flexibility of the particle. At each in-  
184 tersection between a cylinder and a joint a virtual disk is present, so every  
185 joint consists of two disks. These two disks are interconnected with six linear  
186 spring-dampers, positioned in an axisymmetric pattern. The stiffness of the



187 springs determines the bending resistance of the joint and the dampers pre-  
 188 vent the joint from oscillating by dissipating the received energy. The springs  
 189 are not loaded if the two disks coincide. The tensile stiffness of each segment  
 190 of the straw particle is provided by one extra set of a spring and damper.  
 191 They are attached to the two disks belonging to that particular segment.

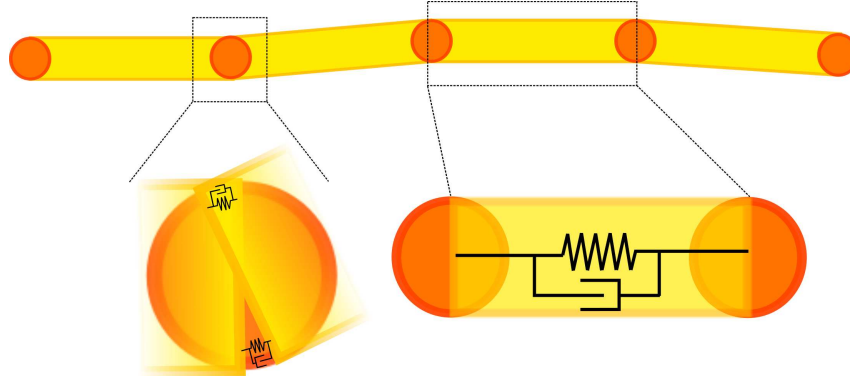


Figure 1: Bendable straw in DEM

### 192 2.3. Calibration of particle properties

193 As the grain particles are conceived as spherical particles, their properties  
 194 can be directly set in DEMeter++. The proposed construction for bendable  
 195 straw on the other hand needs calibration by adjustment of the different  
 196 spring and damper constants in order to obtain a realistic behaviour. As  
 197 most of the mechanical properties of straw have been described in literature  
 198 these can be used here. The ones that are considered relevant for separation  
 199 are summarized together with their source in Table 1. As the mechanical  
 200 properties of straw can vary considerably, parameter ranges are given instead  
 201 of single values. As no values for the radial compressibility of the straw stalks

202 were found in literature, this property was determined experimentally as will  
 203 be described in section 2.3.3.

Table 1: Properties of wheat straw employed in the simulations

Straw properties	minimum	maximum	source
length ( $m$ )	0.7		b
radius ( $m$ )	0.00142	0.001725	c, d, i
wall thickness ( $m$ )	0.000527	0.000957	c, d
density ( $kg/m^3$ )	195.97	226,62	c, d
$E_b$ (GPa)	1.283	5.995	c, d, g
friction (straw-straw) <sup>1</sup>	0.3		h
friction (straw-steel) <sup>2</sup>	0.3		h
tension modulus ( $GPa$ )	4.16	8.22	f
Grain properties	minimum	maximum	source
minor axis diameter ( $m$ )	0.00254	0.0035	a, e
kernel weight ( $kg$ )	0.00003366	0.00005084	a

Sources: a:Mohsenin et al. (1986); b:Beck (1992); c:O'Dogherty et al. (1995);  
 d:Annoussamy et al. (2000); e:Stroshine (2000); f:Wright et al. (2005);  
 g:Tavakoli et al. (2008); h:Sitkei (1986); i:own measurements.

<sup>1</sup> Moisture content 25% W.B.; <sup>2</sup> Moisture content unavailable

### 204 2.3.1. Straw bending stiffness

205 The segmented straw stalks in DEM are calibrated to obtain a realistic  
 206 bending behaviour. Most researchers (e.g. Annoussamy et al. (2000) for  
 207 wheat straw) measured the bending stiffness of wheat straw using a three-

point bending test. Straw stalks are cut into their different internodes and each internode is placed on two rounded supports positioned a distance  $L$  apart and loaded in the middle by a moving support. By measuring the applied force ( $F_b$ ), the outer radius of the stalk ( $R$ ) and the thickness of the stem wall ( $e$ ), the Young's modulus in bending ( $E_b$ ) was calculated for each internode using formulas (3) and (4) with  $I_b$  the second moment of area of a tubular cross-section.

$$I_b = \frac{\pi[R^4 - (R - e)^4]}{4} \quad (3)$$

$$E_b = \frac{\frac{L^3(dF_b/dl)}{48}}{I_b} \quad (4)$$

The Young's moduli obtained from Annoussamy et al. (2000) can now be used to determine the bending stiffness ( $dF_b/dl$ ) of a straw stalk with the same length as the straw stalks employed by Beck (1992) (0.7 m) by performing the reverse calculation. The distance between the supports is then chosen to be 0.667 m, which establishes a 2.5% margin at each side so that the particle is not pushed off the supports during bending. Annoussamy et al. (2000) reported a bending stiffness value for each internode. To obtain a general Young's modulus in bending for the whole straw particle a weighted average is taken with the internode lengths as weights. The result of this averaging can be found in Table 1. Next, Eqs. 3 and 4 are used to derive the bending stiffness  $dF_b/dl$ . To apply the obtained value for the bending stiffness  $dF_b/dl$  to the virtual straw a link has to be established between the linear spring-dampers in between the different segments and the calculated bending stiffness. This is done by successive simulations of the three-point bending test in DEMeter++ (Fig. 2) until the slope of the simulated curve

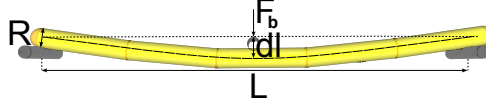


Figure 2: Simulated three-point bending test

coincides with the experimental one. By comparing the slope of the simulated force-displacement curve with the desired  $dF_b/dl$ , the stiffness of the springs and the damping factor of the dampers can be adjusted until the straw shows the right behaviour.

### 2.3.2. Straw elongation stiffness

A second characteristic of the straw stalks that can be controlled in the DEM simulations, is their elongation stiffness. As illustrated in Fig. 1, each segment contains a linear spring-damper mounted in between the disks at each end. This system of parallel springs and dampers is not loaded at the initial straw length. Wright et al. (2005) determined the tension moduli for two varieties of wheat, summarized in Table 1, the smallest value corresponds to the *Westbred 936* variety, the largest one to the *Amidon* variety. As Young's modulus is defined as the ratio of stress and strain (Eq.5) and using the definition of the spring constant  $k = F_e/\Delta L$ , a general spring constant can be calculated easily by combining both equations. As before, the length  $L$  of the straw stalk is set to 0.7 m.

$$E_e = \frac{F_e/A}{\Delta L/L} \quad (5)$$

$$k = \frac{E_e \cdot A}{L} \quad (6)$$

Each segment contains a spring, so the tension modulus of an aggregate straw particle is the result of a series of springs. The spring constant for

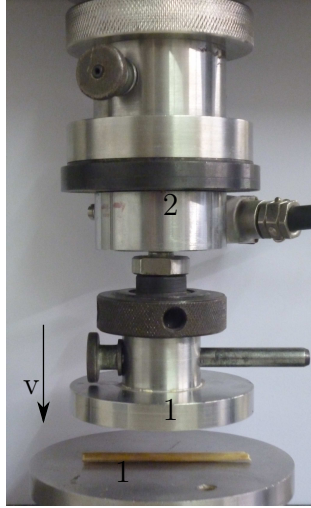


Figure 3: Experimental setup consisting of two approaching flat plates (1) and a load cell (2)

each individual spring should then be the overall spring constant multiplied with the number of segments  $n$ .

### 2.3.3. Straw radial compressibility

The radial compressibility of a straw stalk specifies how the contact between individual straw stalks will occur. The stiffer the material, the larger the impact. In the DEM simulations, this compressibility is contained in the normal contact force model (section 2) as a linear spring-damper. As little is found in literature on single straw compression, a compression experiment was conducted using a Universal Testing System (Type 005.00, *UTS Testsysteme*, Ulm, Germany). The experimental setup is depicted in Fig. 3. A 0.06 m long straw stalk was compressed between two flat plates. The compression force is measured by a load cell (*U1, Hottiger Baldwin Messtechnik*, Darmstadt, Germany) with a range of 200 N. The measurement is repeated

for five straw stalks with 3 internodes each. The resulting force-displacement curve is shown in (Fig. 4). Although the compression process is definitely

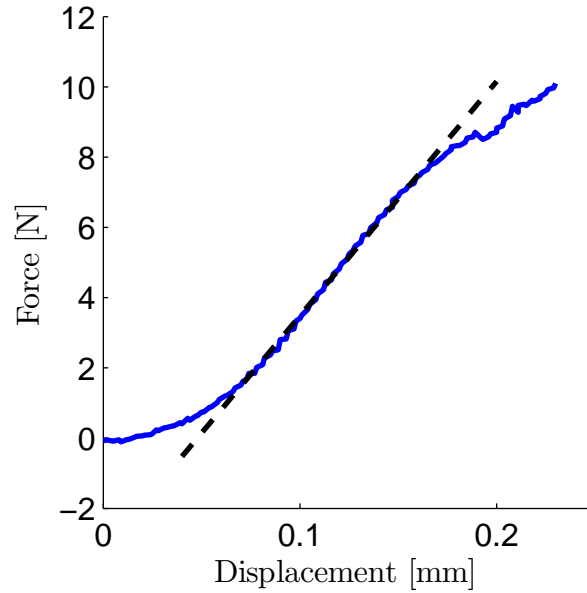


Figure 4: Force-displacement curve

non-linear, which can be seen in Fig. 4, it will be approached in a linear way. Only the first, linear elastic part of the force-displacement curve is taken into account, the deflection in the last part is due to cracking of the particles which is unlikely to happen in the separation experiment, given the low straw area densities employed in the experiments and the relatively high force at which cracking occurs. A straight line is fitted on these measurement points by a least-squares regression. The slope of this trend line is a measure of the compression stiffness. The average of all 15 measured internodes is taken to obtain a realistic value for the compression stiffness of an individual straw stalk of 0.06m. As illustrated in Fig. 5 the stalk forms a system of

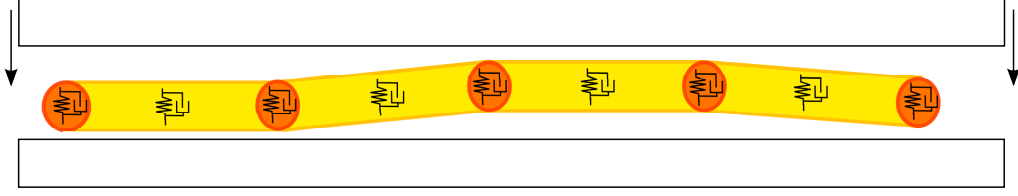


Figure 5: DEM equivalent of the UTS compression test

multiple springs and dampers in parallel in the radial direction so that the compression stiffness obtained from the UTS measurement has to be divided by the sum of the number of spheres and cylinders that constitute a straw particle to obtain the individual spring constants.

#### 2.3.4. Straw friction

The Werner and Haff model is used to model the tangential contact (friction). This model requires a friction coefficient. Sitkei (1986) reported a friction coefficient of straw on galvanized steel of 0.3 and a coefficient for friction between straw particles in function of their moisture coefficient and mutual angle. It is, however, computationally not efficient to include an angle-dependent friction coefficient in the simulations. As simulations with the upper and lower limits of the straw-straw friction values reported by Sitkei (1986) did not result in significant differences, the average value of 0.3 for dry straw was used. Moreover Sitkei (1986) reported that at low moisture contents like the one used in the experiments, the angle dependence diminishes.

### 291 3. Material and methods

#### 292 3.1. Simulated experiment

293 The experiment performed by Beck (1992) has been simulated in DEMe-  
294 ter++. This experiment involves a cubic box with edges of 0.707 m oscillating  
295 sinusoidally in the vertical direction with an amplitude of 0.03 m at a fre-  
296 quency of 4 Hz. At the bottom of the box, a grating holds back the straw,  
297 but allows the grain to pass. Initially the box is filled with straw up to a  
298 certain area density. In the next step, the straw is shaken for 20 seconds  
299 to randomize it. After the randomization a homogeneous layer of grain is  
300 released on top of the shaking straw. A weighing scale underneath the box  
301 records the separated grain fraction in function of time. The experiment has  
302 been repeated for several grain (0.5, 1 and 2  $kg/m^2$ ) and straw area densities  
303 (2,3,4,6 and 7  $kg/m^2$ ) in order to assess their effect on the penetration time.  
304 A significant effect of the straw area density on the separation profiles was  
305 found. Only the straw area density of 2 and 3  $kg/m^2$  could be simulated  
306 within a reasonable time frame. Beck (1992) reported that grain area den-  
307 sity did not have a significant effect in the tested range so 0.5  $kg/m^2$  of grain  
308 was used in the simulations to reduce computational time.

#### 309 3.2. Structure of the simulations

310 Although a real-life separation experiment takes a short time to execute,  
311 the corresponding simulation is computationally expensive due to the high  
312 number of particles involved. The simulation consists of 5 phases. At some  
313 points, the simulation has been changed in comparison to the experiment to  
314 shorten the simulation time. First of all, the ground surface of the box has



315 been reduced by a factor four such that less particles are required to simulate  
 316 the area densities used in the experiments by Beck (1992). The influence of  
 317 this intervention has been verified. Additional changes with reference to the  
 318 experiment will be indicated in the description of the different phases below.

319 *Initialization of the straw layer.* The straw is aligned at the bottom of the  
 320 box in horizontal layers as visualized in Fig. 6. The center lines of the straw  
 321 particles in a layer are 0.01 m apart, horizontal layers are also at 0.01 m  
 322 apart. Subsequently, the joints of each straw particle are given a random  
 323 initial deviation with a maximum of 2.5 mm, because also true straw stems  
 324 are not completely straight and it reduces the number of contacts at the first  
 325 moment that the straw particles touch.

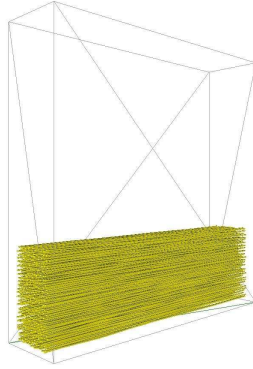


Figure 6: Straw initialization

326 *Randomization of the straw layer.* After the initialization of the straw layers  
 327 the oscillatory motion is started. In this way, the straw is shaken up in  
 328 the box for 2 s (Fig. 7). As in a real experiment the particles experience  
 329 aerodynamic drag, this effect has also been implemented in the simulation.  
 330 The straw randomization was shortened from 20 s in the real experiment to

331 2 s to reduce the simulation time. The time step of discretization was set to  
 332  $6 \mu s$ . Subsequently, the state of all particles is saved.

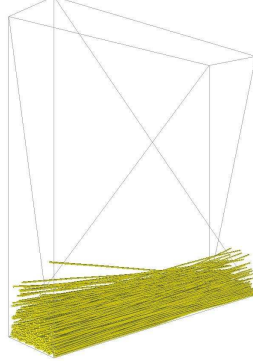


Figure 7: Straw randomization

333 *Grain initialization.* The grain particles ( $0.5 \text{ kg/m}^2$ ) are randomly spread  
 over the area of the box in one layer at the top of the box (Fig. 8).

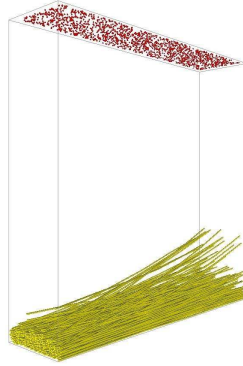


Figure 8: Grain initialization

334

335 *Grain and straw simulation.* After releasing the grain layer above the straw  
 336 mass, the kernels fall into the box, resisted by aerodynamic drag forces.  
 337 After about 0.5 s, the kernels reach the top of the straw mass. The straw

338 restrains the kernels' movement. As the straw is randomized, the particle  
 339 trajectory lengths as well as their velocity on the trajectories differ. When the  
 340 kernels approach the grit floor they pass through freely as there is no contact  
 341 detection between the kernels and the floor of the box (Fig. 9). This differs  
 342 from physical experiments where a grid floor slightly slows down the grain  
 343 kernels. This influence is, however, very limited as stated by Beck (1992)  
 344 because a grid of  $1 \times 1 \text{ cm}^2$  with a wire diameter of 1 mm was employed. Due  
 345 to the large number of contacts between the particles, the time discretization  
 346 had to be reduced from  $6 \mu s$  to  $2 \mu s$  compared to the randomization phase.

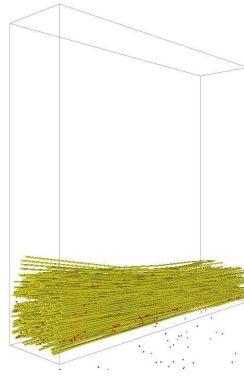


Figure 9: Separation

347 *Post-processing.* In the grain simulation step the grain kernels passing through  
 348 an imaginary plane at 5 cm below the floor of the box are counted. This is  
 349 different from the experiment described by Beck (1992) where the separated  
 350 grain fell onto a weighing scale fixed to the base frame of the setup. Conse-  
 351 quently, the travelling time of a kernel between the grit floor and the weighing  
 352 scale depends on the location of the box at the moment of separation, which  
 353 could introduce small variations between the experimental and simulation

354 results. For the validation, this does not make any difference because the  
355 weighing was done at a much lower frequency than the oscillation of the box.  
356 However, from a modeling point of view it is more useful to have the kernels  
357 travelling a fixed distance before detection as the distance that the kernel is  
358 subjected to aerodynamic drag is constant. Only the plane of detection is  
359 moving, for which can be corrected easily.

360 In post-processing, the cumulative separated fraction is plotted as a func-  
361 tion of time.

362 Summarizing, three simplifications have been introduced in order to make  
363 the simulation computationally feasible. First of all, the box size has been  
364 reduced, the lowest grain area density employed by Beck (1992) has been  
365 employed and the randomization phase has been shortened. These changes  
366 allow to simulate the experiment up to  $3 \text{ kg/m}^2$  of straw within a reasonable  
367 time. Simulating a higher amount of straw particles is currently not feasible  
368 because of too long simulation times due to the fact that the DEMeter++  
369 code has not been parallelized yet.

370 The experimental separation profiles from Beck (1992) start when the  
371 first grain kernel is separated. In order to make a valid comparison between  
372 the simulated curves and the experimental ones, also the origin of the sim-  
373 ulated curves needs to be shifted to the point at which the grain starts to  
374 be separated from the straw mass. It was opted for here to use the time at  
375 which the fifth grain kernel has left the box as this starting point, as the  
376 moment when the first grain kernel is separated was found vary too much.

### 3.3. Experimental design of the simulations

To validate the DEM modeling of grain-straw separation with bendable straw the simulated separation profiles were compared to the profiles measured by Beck (1992) and the separation profiles resulting from the DEM equivalent. However, Beck (1992) did not report the straw and grain properties of the crop used in his experiments. He only reports that the same crop was used in all experiments. Therefore, DEM simulations have been performed for a range of realistic wheat straw and grain properties. The DEM approach is considered valid when the simulated separation profiles for a set of crop properties at all simulated straw area densities match sufficiently well with the profiles reported by Beck (1992). Beck (1992) performed separation experiments at different grain ( $0.5$  to  $2 \text{ kg/m}^2$ ) and straw densities ( $2$  to  $7 \text{ kg/m}^2$ ). However, he reported that the grain area density had no significant effect on the separation profile. Therefore, simulations have only been performed at a grain area density of  $0.5 \text{ kg/m}^2$ . Additionally, the simulations were limited to the  $2 \text{ kg/m}^2$  and  $3 \text{ kg/m}^2$  straw area densities (*SAD*), because the higher straw area densities could not be simulated for all considered crop properties (see below) within a reasonable time frame.

To limit the required number of simulations, only the crop properties which were expected to have a major influence on the separation profile have been included in the design. For the straw particles these properties are mass per unit length ( $ML$ ), radius ( $R$ ) and Young's modulus in bending ( $E_b$ ). To calculate a realistic mass per unit length of the straw stalk two extra parameters have to be known: density and wall thickness ( $e$ ). For the grain kernels only the kernel weight ( $GW$ ) and minor axis radius ( $GR$ ) are selected. The

402 minor axis diameter is chosen as the radius of the spherical grain kernels in  
 403 DEM, ignoring the major axis diameter because the ease of sinking of a par-  
 404 ticle through a particle bed is mainly determined by the smallest diameter  
 405 of the particle. Performing DEM simulations for all possible combinations  
 406 of crop properties (full factorial design) is not computationally feasible due  
 407 to the number of crop properties and the high computational cost per sim-  
 408 ulation (on average 21 days computation time). Therefore, a uniform space  
 409 filling design of crop properties has been constructed in the JMP software  
 410 (*JMP, The SAS Company, Inc., Cary(NC), USA*). This design uniformly  
 411 samples the virtual domain spanned by the crop properties (6) while keeping  
 412 the amount of necessary simulations as low as possible. To obtain a valid  
 413 experimental design, the amount of test points needs to be increased with 10  
 414 for each dimension added (Fang et al., 2000). Thus, the result is a uniform  
 415 design with six dimensions: 4 straw properties and 2 grain properties consist-  
 416 ing of 60 combinations for which DEM simulations had to be performed. The  
 417 same experimental design of crop properties has been used for 2 and 3  $kg/m^2$   
 418 straw area density. Using the techniques for crop calibration, the desired crop  
 419 properties were then translated into simulation parameters. The other crop  
 420 variables which have to be defined in the DEM simulations were kept con-  
 421 stant at the earlier mentioned average values. All 120 simulations have been  
 422 performed in parallel, each on a 2.8 GHz Intel Xeon 5560 (Nehalem) CPU of  
 423 the KU Leuven/UHasselt VIC3 High Performance Computing cluster.

#### 424 3.4. Validation

425 As the separation profiles have an S shape, 10 candidate sigmoidal equa-  
 426 tions have been fitted on the separation profiles and judged by their coefficient

427 of determination and absence of lack-of-fit. The following sigmoidal equation  
 428 was selected for its accuracy and interpretability (Eq. 7):

$$S(t) = (1 - e^{-a \cdot t})^b \quad (7)$$

429 Where  $S(t)$  is the cumulative separated fraction,  $a$  is a measure of the  
 430 slope of the curve,  $b$  shifts the curve to the right. Next, the coefficient of  
 431 determination between the simulated and the measured profiles has been  
 432 computed for each of the 60 separation curves of an area density series.

### 433 3.5. Sensitivity study

434 The DEM simulations for the 60 combinations of crop properties of the  
 435 uniform space filling design at each straw area density provide information on  
 436 the effect of straw area density and crop properties on the grain separation  
 437 profiles. The shape of the simulated separation profile associated with each  
 438 combination of crop properties is described with the  $a$  and  $b$  parameter of  
 439 the sigmoidal function fit. Baader et al. (1969), Beck (1992) and Huisman  
 440 (1978) derived as additional parameter the time to separate a defined fraction  
 441 of the grain mass, because this gives an intuitive indication of the separation  
 442 rate. Usually this fraction is taken as 80%, to get the so-called  $t_{0.8}$  value.  
 443 Evaluating Eq. 7 in  $t = 0.8$ ,  $t_{0.8}$  can be expressed in function of  $a$  and  $b$  as:

$$t_{0.8} = \frac{\ln(1 - \sqrt[b]{0.8})}{a^2} \quad (8)$$

444 Next, an all possible subsets regression has been carried out in the JMP  
 445 software to select the optimal subset of straw area density, crop properties  
 446 and two-factor interactions for a regression model to predict  $a$ ,  $b$  and  $t_{0.8}$ .

447 The subsets that included interaction terms were restricted to exhibit effect  
448 heredity so that an active 2-factor interaction will have at least one of its main  
449 effects also active. The algorithm outputs the best subset in function of the  
450 number of parameters. The optimal number of parameters has been selected  
451 based on Akaike's information criterion ( $AICc$ ), the Bayesian Information  
452 Criterion ( $BIC$ ), Mallow's  $Cp$  and the adjusted  $R^2$ .

## 453 4. Results

### 454 4.1. Validation of the bendable straw approach for grain-straw separation

455 The separation profiles associated with the different combinations of crop  
456 properties generated with the space filling design are displayed in Fig. 10 for  
457 a straw area density of  $2\text{ kg/m}^2$  and in Fig. 11 for  $3\text{ kg/m}^2$ . These 60 different  
458 separation curves show that the influence of crop properties on separation is  
459 substantial, as a large variation is present. Also a 4 Hz modulation due to the  
460 oscillation of the box, is visible on the separation curves. It can be observed  
461 that a number of simulated curves correspond well to the experimental ones.

462  
463 The coefficient of determination ( $R^2$ ) between the simulated and the ex-  
464 perimental data is computed for each of the 60 separation curves of an area  
465 density series. As shown in Table 2, eight different combinations of crop  
466 properties deliver an  $R^2$  value over 0.98 for both straw densities. It is re-  
467 markable that several, apparently unrelated combinations of crop properties  
468 are able to match the experimental separation profiles of Beck (1992). This  
469 indicates again the complex interactions of the crop properties underlying  
470 separation.



Table 2: Crop properties of good fitting DEM simulations of 2 and 3kg/m<sup>2</sup>

	Straw properties			Grain properties		$R^2_{2kg/m^2}$	$R^2_{3kg/m^2}$
	$R$	$ML$	$E_b$	$GR$	$GW$		
	( $10^{-3}m$ )	( $10^{-3}kg/m$ )	( $GPa$ )	( $10^{-3}m$ )	( $10^{-5}kg$ )		
	Range	[1.42,1.73]	[0.85,1.54]	[1.28,6.00]	[2.54,3.50]	[3.30,5.08]	
1	1.69	1.48	3.81	2.93	4.33	0.99	0.98
2	1.59	1.39	3.04	3.16	4.50	0.99	0.99
3	1.65	1.25	2.64	2.62	4.65	0.99	0.99
4	1.63	1.32	4.23	2.73	5.03	0.99	0.99
5	1.61	1.26	3.13	2.69	3.84	0.99	0.99
6	1.63	1.46	2.97	3.22	3.39	0.99	0.99
7	1.69	1.45	5.58	2.74	3.61	0.99	0.99
8	1.60	1.23	1.42	3.06	3.56	0.98	0.99

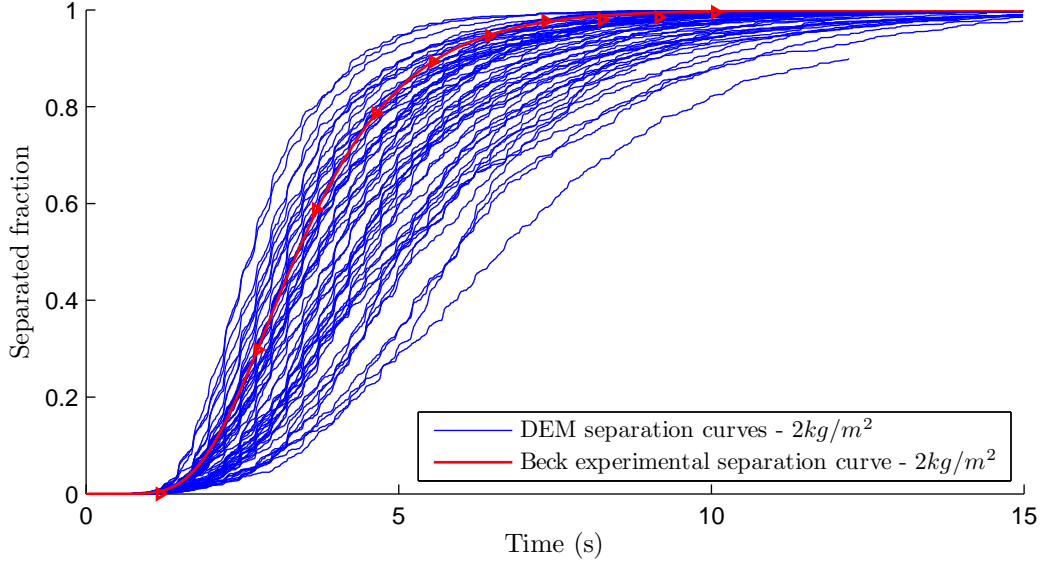


Figure 10: Separation curves resulting from the uniform space filling design for a straw area density of  $2 \text{ kg/m}^2$

471 It can be observed that the simulations matched generally better with the  
 472 higher straw radii ( $R$ ) and the higher straw masses of the experimental design.  
 473 However, a simulation with a high straw mass and high grain diameter can for  
 474 instance result in a similar separation profile to a simulation with a low straw  
 475 mass and a small grain diameter. This suggests that certain crop properties  
 476 have similar or opposite effects on separation. Beck (1992) used one type of  
 477 crop for the experiments. Also here, the same sets of crop properties provide  
 478 successful matches at both straw area densities.

479 It can be concluded from the high coefficients of determination and the  
 480 fact that similar optimal combinations show up in both area densities that the  
 481 experimental separation profiles from Beck (1992) can be reproduced with

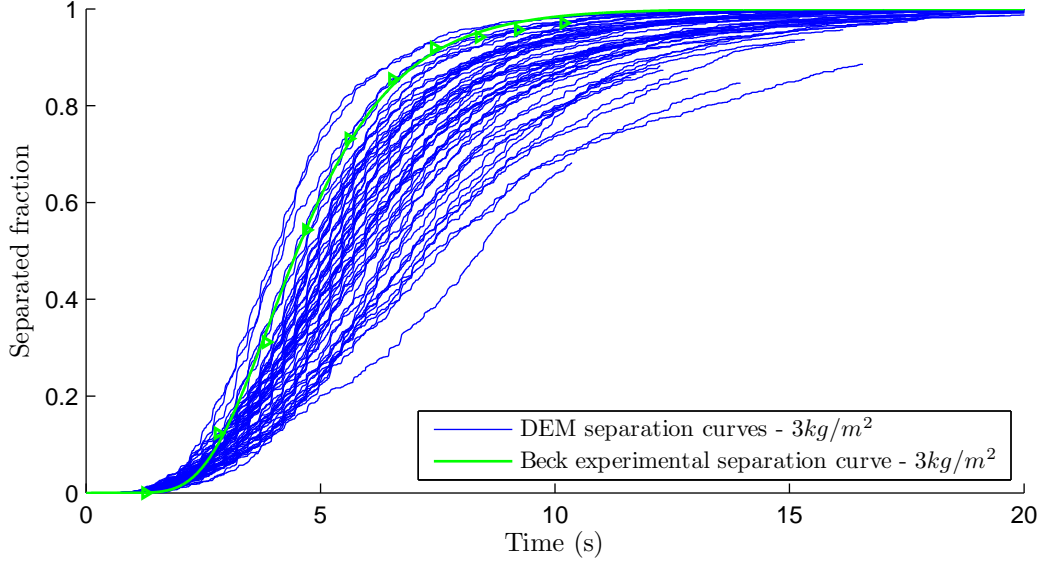


Figure 11: Separation curves resulting from the uniform space filling design for a straw area density of  $3 \text{ kg/m}^2$

482 a DEM simulation using bendable straw particles and grain with realistic  
 483 properties. This endorses DEM as a tool to perform grain-straw separation  
 484 experiments. However, it is not possible to point out one single best set of  
 485 crop properties. This is due to two facts: First, there is a large variation on  
 486 the position of each separation curve due to the randomness in grain kernel  
 487 initialization. Secondly, some crop properties have similar or opposite effects  
 488 on separation, which makes different combinations of realistic crop properties  
 489 feasible. In order to get a better insight in the influence of individual crop  
 490 properties on separation, a sensitivity analysis has been conducted and the  
 491 results are presented in the next section.

#### 4.2. Sensitivity study of particle properties on grain-straw separation

An all possible subsets regression has been performed to find the combination of straw area density (*SAD*) and crop properties that gives the best prediction of the  $a$ ,  $b$  and  $t_{0.8}$  parameters of the sigmoidal fit to the simulated separation profiles. The coefficients of the regression functions for  $a$ ,  $b$  and  $t_{0.8}$  are shown in Table 3. The selected models for  $a$  and  $t_{0.8}$  have a high coefficient of determination (both  $R^2 = 0.80$ ), while the one for  $b$  does not ( $R^2 = 0.44$ ). All effects included in the table are significant ( $P < 0.05$ ). As can be seen, all main effects are active ( $P < 0.05$ ), except grain kernel weight for  $t_{0.8}$ . A limited number of interaction terms is active ( $P < 0.05$ ), but not consistently.

In the range of  $a$  ( $[0.3, 1]$ ) and  $b$  ( $[5, 20]$ ), the gradient of  $t_{0.8}$ , derived from Eq. 8 in the direction of  $a$  varies between  $-50$  and  $-3$  while in the direction of  $b$  it varies from  $0.05$  to  $0.65$ . This means that the effect of  $b$  on the separation rate is small, compared to the effect of  $a$ .

The main effects on  $a$  have the same sign of the effects on  $t_{0.8}$ , which is not the case for the main effects on  $b$ . Due to the predominance of the effect of  $a$  on separation rate and the low correlation with  $b$ , the sensitivity study will be confined to the effect of crop properties on  $a$  and  $t_{0.8}$ . Looking at the observations by Beck (1992), however, it is likely that the effect of  $b$  will become more pronounced at straw area densities higher than the ones simulated in this research. When comparing Fig. 10 to 11, slower separation is noticed at higher straw area density. The increased path length of the grain, with higher probability of dispersion after collision with the straw particles, results in slower separation. This is confirmed by the observations

517 of all previously cited separation studies. The slower separation translates in  
 518 a higher  $b$  value while the increased dispersion flattens the separation profile,  
 519 resulting in a lower  $a$  value. Both effects combine into a higher separation  
 520 time  $t_{0.8}$ . Also the effects of the straw radius  $R$  and the mass per unit length  
 521 of the straw particles  $ML$  have to be viewed in that light. A higher straw  
 522 radius will result in higher straw coverage and likewise in a higher probability  
 523 of collision of the grain with straw particles. This results in a lower  $a$  value, a  
 524 higher  $b$  value and a higher  $t_{0.8}$ . As the simulations have been carried out with  
 525 a fixed straw area density, a higher mass per unit length ( $ML$ ) means that  
 526 the total amount of straw stalks will be lower, thus reducing the thickness of  
 527 the straw layer and speeding up separation. However, if the stalks become  
 528 heavier, they will be thrown less high by the oscillating box, implying that  
 529 the straw mass stays more dense and by this is harder to penetrate, but  
 530 apparently the reduction in the amount of straw particles is more dominant.  
 531 A higher bending stiffness of the straw ( $E$ ) decreases the separation speed.  
 532 It was observed that less bendable straw particles stay more aligned in the  
 533 box. In this way they pack densely and restrain the kernels more. This effect  
 534 is even amplified at higher straw area densities and higher straw radii, as  
 535 indicated by the active interaction of  $E$  and  $SAD$ . Increasing grain radius  
 536 ( $GR$ ) slows down separation. This confirms the observations of Gregory  
 537 and Fedler (1986) and Shandilya (1987). This can possibly be explained by  
 538 the fact that small grain kernels can penetrate much narrower voids. The  
 539 negative sign for the interaction between  $GR$  and  $R$  indicates that the effect  
 540 of grain radius decreases when increasing the straw radius which suggests  
 541 that separation speed depends on the relative size of the particles involved,

542 which is rather intuitive.

#### 543 *4.3. Computational requirements*

544 All simulations have been executed in parallel on a separate processor  
545 core. The average computing time depended on the number of particles in-  
546 volved in the simulation. On average, it took 10 h to compute a second of  
547 simulation for a straw area density of  $2\text{ kg/m}^2$  and 24 h for a straw area  
548 density of  $3\text{ kg/m}^2$ , both with the reduced box size. For  $2\text{ kg/m}^2$  also short-  
549 ened simulations with the full box were done. this took 139 h per second of  
550 simulation, making validations with the full box size computationally infea-  
551 sible. For  $3\text{ kg/m}^2$  no comparison was made. Even though the computations  
552 have been executed on modern CPU's, computational power stands in the  
553 way of a full validation. DEM simulations can be accelerated by running the  
554 simulations on a graphics processing unit (GPU) instead of a central pro-  
555 cessing unit (CPU) or by parallellization of DEMeter++ allowing the use of  
556 multiple cores for one simulation. The new upcoming version of DEMeter++  
557 promises a 20-fold increase of the calculation speed. If applied in parallel, an  
558 additional gain of a factor 6 is expected.

### 559 **5. Conclusions**

560 In this research, grain-straw separation was simulated with spherical grain  
561 particles and segmented straw particles made up of cylindrical and spherical  
562 DEM primitives. Before conclusions were drawn concerning the influence  
563 factors on grain-straw separation, the approach was successfully validated  
564 against the experiments of Beck (1992). A number of candidate sets of crop  
565 properties have been found to fit both the  $2$  and  $3\text{ kg/m}^2$  straw area densities

Table 3: Coefficients of the crop simulation parameters in all possible subsets regression

	a ( $R^2 = 0.80$ )	b ( $R^2 = 0.44$ )	$t_{08}$ ( $R^2 = 0.80$ )
Intercept	1.69	1.50	-6.35
$SAD$	$-1.80 \cdot 10^{-1}$	$6.20 \cdot 10^{-1}$	2.34
$R$	$-4.29 \cdot 10^2$	$6.96 \cdot 10^2$	$4.47 \cdot 10^3$
$ML$	$5.99 \cdot 10^2$	$1.09 \cdot 10^2$	$-6.58 \cdot 10^3$
$E_b$	$-1.55 \cdot 10^{-2}$	$1.50 \cdot 10^{-1}$	$1.31 \cdot 10^{-1}$
$GR$	$-2.14 \cdot 10^2$	$-1.44 \cdot 10^2$	$2.18 \cdot 10^3$
$GW$	$-7.53 \cdot 10^2$	$-1.59 \cdot 10^4$	
$SAD_n \cdot E_{b,n}$	$4.92 \cdot 10^{-2}$		$3.50 \cdot 10^{-1}$
$SAD_n \cdot R_n$		$-7.32 \cdot 10^3$	
$SAD_n \cdot ML_n$		$6.53 \cdot 10^3$	
$SAD_n \cdot GR_n$		$-1.69 \cdot 10^3$	
$R_n \cdot E_{b,n}$	$4.03 \cdot 10^2$	$2.08 \cdot 10^3$	$1.45 \cdot 10^3$
$R_n \cdot GR_n$	$-8.05 \cdot 10^5$		
$R_n \cdot GW_n$	$-5.00 \cdot 10^7$	$-4.75 \cdot 10^8$	
$ML_n \cdot E_{b,n}$	$-1.64 \cdot 10^2$	$-9.02 \cdot 10^2$	

**Legend:**  $SAD_n = SAD - 2.50$ ;  $R_n = R - 1.5710^{-3}$ ;  $ML_n = ML - 1.1710^{-3}$ ;  
 $E_{b,n} = E - 3.64$ ;  $GR_n = GR - 3.0210^{-3}$ ;  $GW_n = GW - 4.2210^{-5}$ ;

566 experiments well. Therefore, it was concluded that different combinations of  
567 crop properties lead to the same separation rate. To find out in which direc-  
568 tion and to which extent the properties influence separation, also a sensitivity  
569 study was done on the available runs. Different correlations for the different  
570 straw area densities were found. All crop properties were found to influence  
571 the separation speed, but the main influences can be related to straw cover-  
572 age and grain kernel diameter. The fastest separation occurs in straw with  
573 thin but stiff (in bending) and heavy stalks. This can be the result of a thick  
574 straw wall in combination with a high wall density. These heavy stalks result  
575 in a low number of stalks needed for a certain straw area density and their  
576 small radius gives a low coverage. A low stiffness in bending gives a high  
577 porosity of the straw mat that is easy to penetrate by the grain kernels. The  
578 smaller the radius of these kernels, the more free paths are available, which  
579 results in a faster separation. The effect of moisture on separation, observed  
580 in several researches, affects the particle interactions leading to separation  
581 indirectly, by its influence on the mechanical particle properties such as the  
582 bending stiffness. The relation between these properties and moisture can  
583 be found in for instance Sitkei (1986) and Mohsenin et al. (1986).

584 The DEM-particles developed in this work can be easily applied in com-  
585 bination with more complex geometries such as the walker section of a con-  
586 ventional combine harvester. As Beck (1992) used unthreshed straw stalks of  
587 which the ears were cut, also the properties of unthreshed stalks have been  
588 used in this publication. To simulate a walker section, however, the mechan-  
589 ical properties of threshed, and thus possibly damaged, straw should be mea-  
590 sured and employed. Besides, also break-up of stalks can occur. Therefore,



591 the size distribution of the particles fed to the walkers, which was considered  
592 as an important factor for separation speed by Huisman (1978) should corre-  
593 spond to the one in the real combine. To simulate additional straw break-up  
594 on the walkers, also breakable bonds between the straw segments should be  
595 implemented in the DEMeter++ software.

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675 **Appendix A. List of abbreviations**

$a$	model parameter
$a_i$	acceleration ( $m/s^2$ )
$A$	particle area ( $m^2$ )
$AIC_c$	Akaike's information criterion
$b$	model parameter
$BIC$	Bayesian information criterion
$C_p$	Mallow's criterion
$dl$	Bending deformation ( $m$ )
$e$	straw wall thickness ( $m$ )
$E_b$	Young's modulus in bending ( $GPa$ )
$E_e$	Young's modulus in tension ( $N/m^2$ )
$F_b$	straw bending force ( $N$ )
$F_c$	contact force ( $N$ )
$F_e$	extension force ( $N$ )
$G, H$	body forces ( $N$ )
$GR$	grain radius ( $m$ )
$GW$	grain weight ( $kg$ )
$I$	inertia tensor ( $kg \cdot m^2$ )
$I_b$	second moment of area ( $m^4$ )
$k$	spring constant ( $N/m$ )
$L$	straw length ( $m$ )
$m$	mass ( $kg$ )
$ML$	straw unit mass ( $g/m$ )
$R$	straw radius ( $m$ )
$r_c$	position of contact w.r.t. center of mass of particle ( $m$ )
$S$	cumulative separated fraction
$t$	time ( $s$ )
$\alpha_i$	rotational acceleration ( $rad/s^2$ )
$\Delta t$	time step ( $s$ )
$t_{0.8}$	time to separate 80% of the grain mass ( $s$ )